

# **Cooperation for innovation and its impact on technological and non-technological innovations: empirical evidence for European SMEs in traditional manufacturing industries**

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## **Abstract**

Drawing on a sample of small and medium-sized enterprises (SMEs) in traditional manufacturing industries from seven EU regions, this study investigates how cooperation with external organizations affects technological (product and process) innovations and non-technological (organizational and marketing) innovations as well as the commercial success of product and process innovations (i.e. innovative sales). Our empirical strategy takes into account that all four types of innovation are potentially complementary. Empirical results suggest that cooperation increases firms' innovativeness and yields substantial commercial benefits. In particular, increasing the number of cooperation partnerships has a positive impact on all measures of innovation performance. We conclude that a portfolio approach to cooperation enhances innovation performance and that innovation support programs should be demand-led.

**Key words:** SMEs; cooperation for innovation; traditional manufacturing; technological and non-technological innovations; cooperation breadth.

## **1. Introduction**

The broad context of this paper is the European Commission's 'key priorities for industrial policy' (European Commission, 2014a, p.2). Innovation has now been joined by reindustrialization and a corresponding emphasis on manufacturing industry embracing not only high-tech sectors but also traditional industries, while continuing to "mainstream" SMEs

(European Commission, 2013 and 2014a).<sup>1</sup> Responding to these priorities, this study utilizes a new survey sample of SMEs from six traditional manufacturing industries in seven EU regions to investigate performance effects of cooperation not only on technological innovations but also on non-technological innovations.

As well as investigating the broad topic of cooperation for innovation, this study has a unique focus on SMEs in traditional manufacturing industries. This addresses substantial gaps in the research literature. Previously in the literature, most studies on the impact of cooperation on innovation performance do not report separate findings for SMEs or for SMEs in specific industries or sectors (not distinguishing, for example, between high-tech, low tech, and traditional manufacturing). Moreover, traditional manufacturing industry is largely neglected in the innovation literature. Yet, SMEs in traditional sectors are currently of first-order importance in EU employment and corresponding policy concerns. Authors (2016) eschew the identification of traditional manufacturing with “low-tech”, instead defining traditional manufacturing industry as a coherent unit of analysis in terms of a number of related characteristics: long established; once a main source of employment, at least at the sub-regional level; in the mature or declining phase of their industry life-cycle, with recent decline typically associated with globalization; relatively labour intensive, hence vulnerable to out-sourcing to other countries; but retaining a capacity for innovation.<sup>2</sup>

The authors also document that traditional industries conceptualised in this manner include the six considered in this study: leather and leather products; ceramics or other non-metallic mineral products; textiles and textile products; mechanical/metallurgy or basic metals and fabricated metal products; automotive or motor vehicles, trailers and semi-trailers; and food products and beverages. In the period 2009-12, these six industries accounted for upwards of 40% of all manufacturing jobs in the seven EU regions considered in this study. Moreover, traditional manufacturing industry retains a significant presence throughout the EU, such that in around half of EU regions, the share of traditional industries in manufacturing employment increased over these 15 years [1995-2009]; and in 78 EU regions, the increase was above 4.5%. Although rather neglected by the innovation literature, traditional manufacturing remains an important source of employment and wealth creation in the developed economies. Our

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<sup>1</sup> Cooperation and networking are found to be used interchangeably in the literature. For instance, Pittaway *et al.* (2004) adopted the definition by Perez and Sanchez (2002: 261), whereby networks are defined as "a firm's set of relationships with other organizations".

<sup>2</sup> For the definition of and extensive documentation on the continued importance of traditional manufacturing industry in most EU regions, see [Authors 2016].

investigation of cooperation for innovation in traditional manufacturing industry is thus not a novelty for its own sake but responds to policy concerns throughout the EU and in the US to better understand and promote traditional sector innovation [Authors 2016].

SMEs innovate differently compared to large firms. Their main hampering factors are associated with the limited human and financial resources. While both SMEs and large firms can explore collaboration with different partners as a complementary source of innovation (Tomlinson, 2010; 2009; Zeng et al., 2010), SMEs might be prone to use external knowledge to a larger degree than large firms (Fernández-Olmos and Ramírez-Alesón 2017; Lasagni, 2012; Van de Vrande et al., 2009). Nowadays, firms can be found cooperating with a diverse network of parties, which enables them to access external knowledge and resources and, in that way, complement their internal innovation activities. The cooperation relationships investigated include: between firms within an enterprise group; with suppliers, customers, and competitors; with other private sector firms (consultants, commercial labs and private R&D institutes); with Higher Education Institutions (HEIs); and with public-sector agencies.

In addition, we investigate the impact of breadth of cooperation on performance, i.e. how the number of different cooperative partners affects innovation performance. Katila and Ahuja (2002) were among the first to examine the effects of the scope and depth of search strategy (i.e. the use of external knowledge sources) on firms' innovation performance. Following this line of investigation, Laursen and Salter (2006) introduced the concept of the breadth and depth of external search strategies and found a curvilinear relationship with innovation performance. Some authors use these concepts to investigate how the breadth and depth of other factors besides the use of external knowledge sources affect innovation performance variously defined as: cooperation for innovation (see Grimpe and Kaiser, 2010; Chen *et al.*, 2011; Ebersberger *et al.*, 2012); innovation objectives (see Leiponen and Helfat, 2010); and types of innovation (see Gronum *et al.*, 2012). We follow the former line of investigation and explore how breadth of cooperation influences SME innovation performance. We cannot investigate the effect of the depth of cooperation due to a lack of information on the intensity of cooperative ties.

Innovation output is measured in two ways: by the introduction of both technological (product and process) and non-technological (organizational and marketing) innovations; and by innovative sales as a proportion of total sales, which measures the commercial success of product and process innovations (see e.g. Aschhoff and Schmidt, 2008; Love *et al.*, 2014). Our modeling strategy takes into account the potentially complementary nature of all four types of innovation. Empirical work on the performance effects of R&D cooperation and, more broadly,

cooperation for innovation, have mostly focused on technological product and process innovations (Pippel, 2014; Sánchez–González, 2014). However, since the European Community Innovation Survey (CIS) was introduced in the early 1990s, the concept of innovation has been extended to take into account non-technological aspects of innovation. This trend resulted in the broad definition of innovation proposed in the *Oslo Manual* (OECD 2005), incorporating non-technological organizational and marketing innovations. Likewise, in the stream of innovation research focused on cooperation, most recent studies have examined how cooperation is associated with non-technological organizational and marketing innovations (see for example: Pippel, 2014; Sánchez–González, 2014). Similarly, Pittaway *et al.* (2004) argue for more research on the influence of cooperation for innovation on technological process and non-technological organizational innovations. In line with these developing concerns in the extant literature, this study investigates whether the impact of cooperation is heterogeneous and conditional on actual types of innovation. This study is one of only a few to investigate the impact of cooperation on non-technological innovations (particularly in the context of SMEs) and is the first of its kind in this stream of research to take into account that technological and non-technological innovations may be associated.

The contribution of this study is two-fold and empirical: first, it addresses issues within this broad topic on which the evidence is still far from conclusive – hence, not compelling from a policy perspective; and, secondly, this topic is investigated for the first time in the context of a sector that is largely neglected by the research literature but nonetheless still of major importance throughout the EU. Consequently, our findings have both managerial and public policy implications, thus contributing to the innovation management literature as well as to the innovation policy literature.

A corollary of our focus on cooperation for innovation by SMEs in traditional manufacturing industries is an implied focus on incremental innovation. In discussing the slowdown of productivity growth across the developed market economies, Nobel Laureate Edmund Phelps (2015, p.56) conjectures that: ‘The plausible explanation of the syndrome ... is a critical loss of indigenous innovation in the established industries like traditional manufacturing and services that was not nearly offset by the innovation that flowered in a few new industries – digital, media and financial.’ While Phelps does not use the term “radical innovation”, his concept of indigenous innovation is similar. The corollary is that in the context of traditional manufacturing industry current innovation is largely incremental. This implication is reinforced by our focus on SMEs, given evidence consistent with the Schumpeter Mark II (1942) hypothesis that large firms are the ones with the greatest propensity to introduce

products with higher degrees of novelty (O'Connor and DeMartino, 2006; Tellis, 2013: 240). Previous studies of cooperation have emphasised that the degree or breadth of cooperation is important in enhancing firms' abilities to develop radical innovation; indeed, that 'collaboration was more frequent among firms pursuing higher level rather than incremental innovations ... because ... firms introducing innovations with a greater degree of novelty are more likely to use a wider range of information sources to develop or improve their products' (Nieto and Santamaria, 2007). The context of the present study is thus not only a contribution in its own right, but entails the additional contribution of investigating cooperation for innovation where innovation is overwhelmingly incremental.

This study is organized as follows: the next section discusses theory and evidence on cooperation for innovation and its impact on firms' innovation performance, particularly in the context of SMEs. The third section on methodology reviews the database used in the study and specifies the model. The fourth section presents and discusses the empirical results. Finally, we present conclusions as well as implications for policy makers and managers.

## **2. Theoretical framework**

The proposed benefits of cooperation on firms' innovation activities are extensive: risk pooling and cost sharing; shortening of the innovation process; fast commercialization of products; obtaining access to complementary and/or similar resources; and access to external knowledge (Hagedoorn, 1993; Belderbos *et al.*, 2004b; Pittaway *et al.*, 2004; Aristei *et al.*, 2016). Theoretical insights into the motivation for establishing and maintaining cooperative relationships are provided by transaction cost economics (Williamson, 1985) and the resource-based view of the firm (Barney, 1991).

Transaction cost economics suggests that the motivation is associated with gaining access to similar resources, whereby internal and external knowledge are treated as substitutes (Santamaria *et al.*, 2009; Vega-Juado *et al.*, 2009). That is, the firm is seen as a substitute for the market, whereby the choice between external procurement and internal production (i.e. the "make or buy" decision) is influenced by minimizing transaction costs. By exploiting similar resources, firms can achieve economies of scale, experience and risk diversification (Hagedoorn, 1993; Arranz and de Arroyabe, 2008). With respect to the actual type of cooperative partners that are conducive to the combining of similar resources, Miotti and Sachwald (2003) report that cooperation between competitors is prominent in this case.

In contrast, the resource-based view of the firm proposes that the motivation behind cooperating for innovation is to gain access to complementary resources (Miotti and Sachwald, 2003; Arranz and de Arroyabe, 2008; De Faria *et al.*, 2010). In relation to cooperative partners, the literature suggests that vertical cooperation (with customers and suppliers) is aimed at utilizing complementary resources. As a result of this, vertical cooperation is also termed symbiotic or differentiated cooperation (Arranz and de Arroyabe, 2008). Besides vertical cooperation, Miotti and Sachwald (2003) found that cooperation with universities is targeted at pooling complementary resources.

Consistent with the resource-based emphasis on firms' capabilities, the concept of absorptive capacity likewise advances the complementarity of internal and external innovation sources (Cohen and Levinthal, 1990). Considered to be necessary for exploring and exploiting external knowledge, firms' internal innovation capacity (i.e. absorptive capacity) is usually proxied by the presence within firms of R&D departments and qualified R&D personnel (Miotti and Sachwald, 2003; De Faria *et al.*, 2010; Huang and Yu, 2011; Love *et al.*, 2014; Cozza and Zanfei, 2016). However, because SME innovation is not captured by formal R&D measures (Santarelli and Sterlacchini, 1990; Ortega–Argilés *et al.*, 2009; Raymond and St–Pierre, 2010) - indeed, SMEs more often conduct informal R&D activities (Kleinknecht *et al.*, 2002) - and because this applies in particular to SMEs in traditional manufacturing industry [Authors 2016], in this study we construct a more direct indicator of firms' absorptive capacity (see below, “model specification”). Next, we review the particular benefits of cooperation with a variety of cooperative partners.

Research interest in cooperation with suppliers can be traced back to the 1980s with Japanese car and electronics manufacturers' successes and was closely associated with the relationships between these firms and their suppliers (Sako, 1994; Liker *et al.*, 1996). Amongst rationales for such cooperation, firms may manage to reduce their risks and mistakes in the design of technological products and processes (Fujimoto *et al.*, 1996; Nishiguchi and Ikeda, 1996). Pippel (2014) suggests that the main incentives for firms to cooperate on technological innovations apply also to non-technological organizational and marketing innovations. While cooperation with customers can be of primary relevance for marketing innovations, cooperation with suppliers could be more focused on organizational innovations.

As well as collaboration with suppliers, similarly positive outcomes may arise from close cooperation between firms and their customers (Fitjar and Rodriguez–Pose, 2013). Accessing customer knowledge may be beneficial for firms' innovativeness. This cooperative tie is particularly valuable in the context of new technologies and products (Tether, 2002;

Bogers *et al.*, 2010) and may be of help in improving existing designs (Shaw, 1994) and in inventing new products or applications. Conversely, the dependence on customer knowledge alone may force producer firms to search for new solutions along more established pathways instead of pursuing new or even radical innovations (Laursen, 2011). However, empirical findings confirm that vertical cooperation with customers and suppliers plays a distinct role in the innovation process, particularly amongst SMEs (De Propriis, 2002; Zeng *et al.*, 2010). This joint development of a product between firms and customers is said to improve market share and product credibility (Tidd and Trehwella, 1997; Tether, 2002) and potentially reduce risks associated with the introduction of a new product to the marketplace (Gemünden *et al.*, 1992; Tether, 2002). Concerning non-technological innovations, cooperation with customers is particularly relevant for marketing innovations (Pippel, 2014; Sánchez–González, 2014). Customers' needs and preferences may also significantly contribute to the introduction of organizational innovations, particularly those focusing on firms' external relations.

Horizontal cooperation with competitors is most frequently found in high technology sectors (Mariti and Smiley, 1983) and often sought as a cost and/or risk reduction strategy. By its very nature it is regarded as a potentially precarious alliance due to the possibility of anticompetitive behavior by the cooperating (sic) firms (Tether, 2002). However, such cooperative alliances may have common problems for which they seek solutions and thus avoid potential areas of market rivalry (Tether, 2002). Regarding non-technological innovations, cooperation with competitors may allow firms to realize and adopt successful organizational structures from their rivals (Pippel, 2014). In addition, firms can develop and implement joint pricing and promotion strategies, or, if cooperating in designing new products, firms can engage in a common marketing strategy for a jointly developed new product (Pippel, 2014; Sánchez–González, 2014). However, all potential pitfalls of cooperating with competitors on technological innovations, such as opportunistic behavior and restrictive knowledge sharing, can arise in cooperating on non-technological innovations (Pippel, 2014; Sánchez–González, 2014).

Firms that cooperate with private sector institutions, experts and consultants may not only seek to manage costs but also to pursue the possibility of shared experiences on innovation, helping the firm to pinpoint and specify its exact needs in innovation, contributing ideas for new needs and solutions (Bessant and Rush, 1995) and offering opportunities to bring outside perspectives into the company (Bruce and Morris, 1998). Furthermore, the role of consultants in undertaking organizational and marketing innovations is derived from their potentially broad knowledge base. Namely, consultants can provide an extensive and expert

knowledge in many areas relevant for introducing non-technological innovations (Pippel, 2014; Sánchez–González, 2014).

Seeking external cooperation with HEIs and other public-sector knowledge providers normally entails little to no commercial or market risk (Cassiman and Veuglers, 2002). It is aimed at knowledge development (Miotti and Sachwald, 2003) via access to academic expertise (Azagra–Caro *et al.*, 2006) to inform both technological and non-technological innovation (e.g. new marketing information; Cohen *et al.*, 2002) as well as at reducing costs (e.g. by securing funds for research; Fontana *et al.* 2006) and/or risks. Concerning non-technological innovations, cooperating with HEIs and public research institutes can foster the introduction of innovations that are radical, rather than incremental in nature (Pippel, 2014), given that their main focus is on conducting basic research and providing a heterogeneous knowledge base (Miotti and Sachwald, 2003). Moreover, universities can suggest improvements in firms' organizational structure and management and provide training and knowledge transfer to firms' employees (Sánchez–González, 2014).

The main advantage of cooperating with firms within the same enterprise group is substantially reduced risk of opportunistic behavior. Firms can cooperate with other firms in the same group on organizational innovations as well as on marketing innovations, such as those related to pricing and marketing strategies (Pippel, 2014).

Collaborating with a range of different partners (i.e. breadth of collaboration) might have a positive innovation effect up to a certain point, after which the returns become negative due to over-search (Laursen and Salter, 2006). Following Koput (1997), there are three potential problems leading to over-search. First, “the absorptive capacity problem” might arise, whereby firms might be overwhelmed with too many innovative ideas. In this case, the inverted-U shaped relationship between the collaboration breadth and innovation performance would suggest that externalizing absorptive capacity initially has a positive effect, which turns into a negative effect when the limit of absorptive capacity is reached (Authors, 2017). Second, some innovative ideas might not be fully exploited because they come at the wrong time (“the timing problem”). Third, “the attention allocation problem” might occur, in which case managers are struggling to dedicate enough time and effort to too many innovative ideas. In the case of cooperation, “the attention allocation problem” would arise because cooperation with different partners is demanding on management attention and once a firm reaches a certain number of collaborative partners, managers do not have enough time and effort to dedicate to additional collaborative partners.



## **2.1. Empirical evidence on the impact of cooperation on firm performance**

Extending the division suggested by De Faria *et al.* (2010) and Un *et al.* (2010), we note that empirical studies in the R&D and innovation cooperation literature can be divided into several categories: i) determinants of R&D and innovation cooperation (e.g. Miotti and Sachwald, 2003; Belderbos *et al.*, 2004a; Arranz and de Arroyabe, 2008; López, 2008); ii) the effect of knowledge spillovers on cooperation (e.g. Cassiman and Veugelers, 2002; Chun and Mun, 2012); iii) the impact of cooperation on innovation performance (e.g. Zeng *et al.*, 2010; Lasagni, 2012; Tomlinson and Fai, 2013; Pippel, 2014; Sánchez–González, 2014); and iv) the impact of cooperation on firm performance (e.g. Belderbos *et al.*, 2004b; Faems *et al.*, 2010; Lasagni, 2012; Zeng *et al.*, 2010). The focus of this research is on the third research strand. Yet, the empirical findings on both the innovation and performance effects of cooperation are ambiguous (Belderbos *et al.*, 2004b) and far from establishing a set of “stylized facts”. Moreover, coverage by type of firm and sector is not yet comprehensive. Nonetheless, a generic conclusion can be derived from the literature; namely, that a portfolio approach to cooperation for innovation is adopted by many firms (Faems *et al.*, 2010) and that different cooperative partners have heterogeneous effects on firms' innovation performance.

Tomlinson and Fai (2013) found that, in the UK, SME cooperation with competitors is insignificant for both forms of technological innovation, cooperation with customers marginally increases the probability of product innovation, and cooperation with suppliers yields a highly significant positive impact on both product and process innovations. These conclusions partially confirm previous findings that cooperating with customers and suppliers enhances product and process innovations (see Reichstein and Salter, 2006; Kaminski *et al.*, 2008; Nieto and Santamaria, 2010).

Comparing inter-firm cooperation with other forms of cooperation, Zeng *et al.* (2010) report that cooperation with customers and suppliers has a larger positive impact on the innovation performance of Chinese SMEs than does cooperation with government agencies, universities and research institutes. Similar results are found in Nieto and Santamaria (2010) for Spanish SMEs. However, some studies indicate an increasing importance of research organizations in firms' innovation activities. For instance, Lasagni (2012), analyzing a sample of SMEs from six European countries, reports that both inter-firm cooperation with suppliers and customers and cooperation with research organizations have equally significant impacts on product innovation.

In the absence of an established set of “stylized facts” on cooperation and innovation, particularly regarding SMEs in traditional manufacturing, we prefer to pose research questions (RQs) than hypotheses. To explore the impact of cooperation on technological innovations in the context of SMEs in traditional manufacturing industries, we formulate the following:

RQ1: How does cooperation with different partners (firms in the same enterprise group, customers, suppliers, competitors, private-sector knowledge providers, HEIs and public-sector agencies) affect technological – product and process – innovations?

Empirical studies on the impact of cooperation on non-technological innovations are even more scarce (Pippel, 2014; Sánchez–González, 2014). Moreover, to our knowledge, no study explores this issue for SMEs. Sánchez–González (2014) reports positive effects of cooperative ties with suppliers, customers, competitors, experts and universities on both organizational and marketing innovations. Conversely, Pippel (2014) emphasizes performance heterogeneity with respect to the various cooperative partners: cooperative relations with suppliers, consultants, universities and other firms within an enterprise group all positively affect both organizational and marketing innovations; yet cooperation with customers increases the probability of introducing organizational innovation without any effect on marketing innovation. Finally, cooperation with government research institutes and competitors do not affect non-technological innovation performance. Therefore, the limited available empirical evidence on the impact of cooperation on non-technological innovations in the context of SMEs lead us to formulate:

RQ2: How does cooperation with different partners (firms in the same enterprise group, customers, suppliers, competitors, private-sector knowledge providers, HEIs and public-sector agencies) affect non-technological – organizational and marketing – innovations?

Besides technological and non-technological aspects of innovation as measures of innovation performance in our study, we also use innovative sales to measure the commercial success of innovation. This is line with the argument in Fernández-Olmos and Ramírez-Alesón, 2017, p.18): “Despite the extensive evidence on the importance of technology collaboration networks (TCNs), many researchers emphasize that our knowledge on the effect of TCNs on the economic success of innovation activities are still limited and ambiguous in the area of SMEs.” Innovative sales is the most frequently used measure of innovation output (Love, Roper, and Vahter 2014). Innovative sales is a direct innovation output, while patents measure an intermediate innovation output (Clarysse, Wright, and Mustare 2009; Pakes and Griliches

1980). Negassi (2004) notes that innovative sales (as a turnover-based measure) could be more appropriate than the technological aspects of innovation (i.e., introduction of product and process innovation) in capturing the effect of non-R&D innovation inputs which, we can assume, are particularly pertinent to SMEs in traditional manufacturing industries (Authors, 2017). Based on this argument, we formulate:

RQ3: How does cooperation with different partners (firms in the same enterprise group, customers, suppliers, competitors, private-sector knowledge providers, HEIs and public-sector agencies) affect innovative sales as a measure of the commercial success of innovation?

Concerning cooperation breadth, empirical evidence uniformly suggests its positive effect on innovation performance (see Chen et al. 2011; Ebersberger and Herstad, 2011; Grimpe and Kaiser, 2010; Love et al., 2014). However, findings on the conjectured curvilinear (inverted U-shape) relationship between the cooperation breadth and innovation performance are mixed. Chen et al. (2011) investigate the curvilinearity in Chinese firms divided into those using STI (science, technology and innovation) and DUI (doing, using and interacting) innovation modes. While the curvilinear relationship is found with regards to the STI mode, there is no evidence in the case of DUI mode. Similarly, Love et al. (2014) found no evidence of a curvilinear relationship between cooperation breadth and innovative sales in Irish firms. In addition, Ebersberger and Herstad (2011) found no evidence of a curvilinear relationship between cooperation breadth and new-to-the market product innovation in Norwegian SMEs. Therefore, few studies explore the curvilinearity of cooperation breadth in the context of SMEs and none of them use measures of innovation performance other than product innovation and innovative sales. This led us to formulate our final research question:

RQ4: Is there a non-linear (inverted-U shaped) relationship between the breath of cooperation and innovation performance?

### **3. Methodology**

#### **3.1. Sample**

The dataset used in the analysis covers the period 2005-2009 and was gathered in 2010 within the GPrix project (full project title: Good Practices in Innovation Support Measures for SMEs: facilitating transition from the traditional to the knowledge economy) commissioned by the European Commission's DG-Research. The purpose of the project was to assess a set of

regional innovation support measures in a representative set of European regions characterized by a large number of SMEs from traditional sectors. Besides questions concerning public support for innovation, the survey questionnaire included questions on the firm's size, revenue, and innovation activities in addition to their cooperative ties. The definition of SMEs adopted in the project is in accordance to the European Commission (2008) guidelines, whereby small firms employ fewer than 50 employees, while medium-sized firms have between 50 and 250 employees. The sample includes 312 SMEs from seven EU regions noted for concentrations of traditional manufacturing industry: West Midlands (UK); North Brabant (Netherlands); Saxony-Anhalt (Germany); Emilia-Romagna (Italy); Comunidad Valenciana (Spain); North/Central (Portugal); and Limousin (France). Traditional industries included in the sample are: leather and leather products; ceramics or other non-metallic mineral products; textiles and textile products; mechanical/metallurgy or basic metals and fabricated metal products; automotive or motor vehicles, trailers and semi-trailers; and food products and beverages.<sup>3</sup>

The survey questionnaire was piloted and amended in response to findings from interviews conducted to gain feedback on the validity of both each individual question and the overall design (paying particular attention to its ease of completion within about 20 minutes). These preparations notwithstanding, project members anticipated the practical difficulty—arising from previous experience—that it would be difficult to obtain large numbers of questionnaire responses from SMEs in traditional sectors. Interview evidence gathered in the course of the project yielded several predictable findings in this regard, including: cultural barriers between owners and managers and the world of (social science) research;<sup>4</sup> owners and managers being too busy, typically having nobody to whom to delegate non-essentials, which include completing questionnaires; and that SME owners and managers dislike paperwork, including questionnaires (we found that even trade associations – organisations that SMEs have chosen to join – find it difficult to obtain information from their own SME members). Accordingly, to obtain responses across both the target regions and the target sectors, a two-fold approach was implemented by project partners. First, to align the sample frame as closely as possible with the target population in each region-industry we used, wherever possible, lists of SMEs

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<sup>3</sup> Details on how the sample was obtained in different regions are available in: GPrix Deliverable 3.3, p.20; and Deliverable 1.7, especially pp.10-14. Deliverable 1.7, pp.16-54, also reports exhaustive descriptive statistics on the sample. These documents are available from: <http://business.staffs.ac.uk/gprix/en/index.htm/> (click on the “Reports” tab).

<sup>4</sup> Interview evidence from the project's case studies indicated that most SMEs in the targeted sectors have not had any contacts with universities, with most managers and owners not having participated in higher education themselves.

provided by general business or industry associations to approach firms by e-mail and web sites or, where this was the only alternative, by post. In addition, we provided an “incentive” for all respondents (a prize draw for one of five £100 vouchers for either a top-class restaurant or a department store).

Our most optimistic estimate suggests a response rate from the targeted SMEs of no higher than 3%. Yet the use of our findings to inform policy depends on their external validity. We do not claim that our SME sample is representative of all SMEs in traditional manufacturing industry. Yet, even if a representative sample were to have been feasible, we argue that it would not necessarily have been useful from a policy perspective. Penrose’s (1959, p.7) classic *The Theory of the Growth of the Firm* addressed a similar issue: “Many firms do not grow, and for a variety of reasons. . . I am not concerned with such firms, for I am only concerned with. . . those firms that do grow.” By analogy, policy makers are concerned to encourage innovative or potentially innovative SMEs to more fully exploit their innovative potential. In the introductory section above, we identify as one of the characteristics of traditional manufacturing industries the retention of “a capacity for innovation” and suggest that this characteristic creates potential for public policy to promote innovation in these industries. Correspondingly, our sample firms are overwhelmingly recent innovators (and the rest are at least sufficiently oriented toward innovation to engage with an innovation survey). As long as such firms are a priority for policy makers, then it is valid to use our results to inform policy (Authors, 2016).

### **3.2. Data**

The sample of 312 SMEs is dominated by innovating firms, as almost all firms (94%) had engaged in innovative activities by introducing some type of technological (product and process) and/or non-technological (organizational and marketing) innovations (for definitions, see the *Oslo Manual*, OECD, 2005). Moreover, the sample includes SMEs from seven EU regions and mainly (80%) belonging to one of six manufacturing industries strongly represented in these regions.

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Descriptive statistics are presented in Table 1. The largest number of firms introduced process and product innovations (83% and 81% respectively). In addition, more than half engaged in non-technological innovations (68% in organizational innovation and 61% in marketing innovation). The modal firm in the sample had 36 employees. Slightly more than one fifth (23%) of firms had experienced “very strong” competitive pressure. On average, the surveyed SMEs exported 20 percent of their sales. Slightly more than a third (36%) of firms invested more resources in innovation in 2009 than in 2005. With respect to firms' innovation capabilities in 2005, the largest number of firms (26%) self-reported above average or leading capabilities in product innovation, whereas the smallest number (13%) reported above average or leading capabilities in organizational innovation. Regarding cooperation partners, the largest number of firms stated that they engaged in vertical cooperation (34% of firms cooperated with customers and 32% with suppliers), followed by cooperation with universities and HEIs (31%), with private sector (consultants, commercial labs and private R&D institutes) (24%) and with public sector (government institutions and public research centers) (21%). Although the literature suggests that mostly large firms tend to cooperate with government labs and HEIs (Mohnen and Hoareau, 2003; Lasagni, 2012), while both SMEs and large firms focus their cooperative efforts on vertical cooperation along the supply chain (Laursen and Salter, 2004; Lasagni, 2012), SMEs in our sample tend to cooperate with HEIs to almost the same degree as with customers and suppliers and to a greater extent than with public sector institutions (31% compared to 21%). Conversely, only a small number of firms stated they engaged in horizontal cooperation with their competitors (9%). Finally, regarding the breadth of cooperation, on average, firms cooperate with two cooperative partners, while there are no firms that cooperate with all seven potential partners.

As data were self-reported, common method variance, arising from the measurement method, could bias the estimates due to systematic measurement error (Podsakoff and Organ, 1986). To check internal validity of the data, we conducted the Harmon's one-factor test (Podsakoff and Organ, 1986). The test encompasses an explanatory factor analysis of all independent variables by using unrotated principle component factor analysis. When the common method bias is unlikely to occur, the first unrotated factor (i.e. factor with the largest share of variance) should account for less than 50% of the total variation in other explanatory variables within the model. In our model, the first factor accounts for around 18% of total variation, which suggests that the common method bias raises no great concern in our model (for a recent application, see Love *et al.* 2014).

### 3.3. Model specification

The four dependent variables in the multivariate probit model are binary indicators measuring firms' engagement in technological and non-technological innovations: the dependent variable *Product innovation* is equal to 1 if the firm introduced any new or significantly improved goods and services in the period 2005-2009 (zero otherwise)<sup>5</sup>; *Process innovation* is equal to 1 if the firm implemented a new or significantly improved production process, distribution method, or support activity for its goods or services (zero otherwise); *Organizational innovation* is equal to 1 if the firm introduced new business practices for organizing procedures, new methods of organizing work responsibilities and decision making or new methods of organizing external relations with other firms or public institutions (zero otherwise); and *Marketing innovation* is equal to 1 if the firm introduced significant changes to the design or packaging of a good or service, new media or techniques for product promotion, new methods for sales channels or new methods of pricing goods or services (zero otherwise).

In addition, we separately investigate the impact of cooperation for innovation on innovative sales measured as the share of total sales accounted for by sales arising from new products and/or processes introduced since 2005. Negassi (2004) suggests that innovative sales (as a turnover-based measure) could be more appropriate than the technological aspects of innovation (i.e. introduction of product and process innovation) in capturing the effect of non-R&D innovation inputs which, we can assume, are pertinent to SMEs in traditional manufacturing sectors. Moreover, Love *et al.* (2014) note that innovative sales is the most frequently used measure of innovation output. The variable *Innovative sales* is a categorical variable: = 1 when innovative sales is equal to 0 percent; =2 when innovative sales ranges from 1 percent to 5 percent; =3 from 6 percent to 10 percent; =4 from 11 percent to 15 percent; =5 from 16 percent to 25 percent; =6 from 26 percent to 50 percent; and =7 when innovative sales are more than 50 percent of total sales.

The explanatory variables of interest measure firms' cooperation activities as dichotomous variables equal to 1 if the firm cooperates with the following potential partners (and zero otherwise): within group (*Coop\_within\_group*); suppliers (*Coop\_suppliers*); customers (*Coop\_customers*); competitors (*Coop\_competitors*); consultants, commercial labs, and private R&D institutes (*Coop\_private sector*); HEIs (*Coop\_HEIs*); and government institutions and public research centers (*Coop\_public sector*). Moreover, to capture the breadth

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<sup>5</sup> The survey questionnaire does not include a question on whether firms introduce radical or incremental innovations, thus we are not able to distinguish between these two types of product innovation.

of cooperation and to explore its relationship with firms' innovation performance, we construct the variable *Breadth*, which is equal to the number of cooperative relationships. That is, the variable is equal to zero if the firm does not cooperate for innovation with any of the seven potential partners, and is equal to seven if the firm cooperates with all of the potential partners (Cronbach's alpha coefficient = 0.61). Looking at Table 1, we see that none of the surveyed firms cooperates with all seven cooperative partners (the maximum value of *Breadth* variable is six). Finally, the variable *Breadth* is squared (*Breadth\_sq*), to enable us to test whether the relationship between the breadth of cooperation and innovative performance is curvilinear (taking an inverted U shape).

Control variables include a continuous variable (*Size*) to account for the heterogeneity of SMEs. We model exporting activities (*Export*) as a continuous variable measuring the share of total sales sold abroad in 2009. Exporting firms might be more innovative than their counterparts, as international competition creates more pressure on firms to innovate (Nieto and Santamaria, 2007; Belderbos *et al.*, 2015). In addition, the model includes a variable measuring competitive pressure (*Competition*) (see e.g. Huang and Yu, 2011), which is equal to 1 if the firms responded 'Very strong' to the question: "How would you judge the competition in your main market(s)?", and zero otherwise. The theoretical industrial organization literature predicts that higher competitive pressure negatively affects innovation, because it reduces monopoly rent generated by innovating firms (Aghion *et al.*, 2005).

Following Blundell *et al.* (1995), our models include firm-level "quasi fixed effects" (or initial conditions). These initial conditions control for firms' permanent innovation capabilities reflected in the pre-sample history of innovation outputs, i.e. firms' innovative capacity with respect to technological and non-technological innovations at the beginning of the period covered by the survey (see also Hagedoorn and Wang, 2012). By controlling for past innovative capacity, we take into account firms' absorptive capacity (see e.g. Miotti and Sachwald, 2003). These effects are modeled by the following variables:

- the dummy variable that measures the resources invested in innovation in 2005 relative to 2009 (*Resources*) (DV = 1 if the firm's response to the question "Five years ago did you devote?" was 'Fewer resources to innovation'; = 0 if 'About the same' or 'More');
- dummy variables measuring the firms' innovation capacities for introducing product/process/organizational/marketing innovations within the industry in 2005 (respectively *Capacity\_product*, *Capacity\_process*, *Capacity\_org* and *Capacity\_marketing*) (DV = 1 for 'Above average' and 'Leading'; = 0 for 'Average' and 'Lagging');



Finally, to control for industry heterogeneity, sectorial dummy variables were included for all six industries of interest: automotive; ceramics; leather; metallurgy; textile; and food processing. The base category is other manufacturing industries. In addition, the model includes six country dummy variables for Germany, Italy, France, Portugal, Spain and the Netherlands (with the United Kingdom being the base category).

### **3.4. Empirical strategy**

Cooperation for innovation can influence innovation output, which our survey measures in two ways: first, by the introduction of product, process, organizational and marketing innovations; and, second, by the proportion of sales due to product and process innovations (innovative sales). To date, both theoretical and empirical research in the innovation literature has been almost exclusively focused on technological product and process innovations, although Schumpeter (1947) had earlier identified other non-technological forms of innovation (such as, organizational innovation and opening up of new markets) (Kaivo-oja, 2009: 206; Pippel, 2014). Moreover, Schumpeter suggested a positive correlation between product and process innovations, which has been confirmed in recent empirical studies (see e.g. Miravete and Pernías, 2006; Martinez-Ros and Labeaga, 2009; Doran, 2012). In contrast, few studies explore whether technological and non-technological innovations are interrelated and, if so, how. To investigate this possibility, we use a multivariate probit model that allows all types of innovation to be related (Schmidt and Rammer, 2007; Pippel, 2014).

The underlying assumption of multivariate probit is similar to the seemingly unrelated regression (SUR) approach; in our model, firms may engage simultaneously in each of four innovation outcomes, which are associated both by common observed and, potentially, by common unobserved determinants. In a similar vein, when analyzing the impact of technological collaboration on product and process innovations, Nieto and Santamaría (2010) apply a bivariate probit model and find that product and process innovations are dependent on each other. Concerning non-technological innovations, Sánchez-González (2014) also utilize a bivariate probit model to investigate the effects of cooperation on organizational and marketing innovations, and the results reveal that these types of innovation are also correlated. In our analysis, we combine arguments from these two streams of research and, following

Schmidt and Rammer (2007), investigate the hypothesis that all four types of innovation are correlated. This approach most closely builds upon Doran (2012) who, in a sample of Irish firms, explored whether product, process and organizational innovations are substitutes or complementary. His study reports either a complementary relationship between these three types of innovations or no relationship, and conversely finds no evidence of substitutability between different forms of innovation.

#### 4. Empirical results and discussion

The correlation matrix showing the Pearson correlation coefficients among the independent variables is presented in Table 2. According to Hair (2005, p.351), correlation coefficients below 0.40 suggests a small strength of association between variables, between 0.40 and 0.70 a moderate association, and above 0.70 high association. Therefore, the correlations shown in Table 2 are overall small to moderate. To assess potential problems with multicollinearity, variance inflation factors (VIFs) and conditioning indices were calculated for all four models shown in Tables 3 and 5. The maximum VIF is 2.15, which is substantially lower than the conservative cut-off of 10 (Fernández-Olmos and Ramírez-Alesón, 2017). Likewise, the maximum conditioning index is 13.45, which is below the cut-off value of 20 (Greene, 2012, p. 90). These results suggest that the regression estimates are not biased by the presence of multicollinearity.

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Insert Table 2 about here

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Insert Table 3 about here

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The estimation of the multivariate probit model with individual cooperative partners (Model 1) is presented in Table 3.<sup>6</sup> Concerning the impact of cooperative relationships on

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<sup>6</sup> Following Cappellari and Jenkins (2003), in the case of a small sample size, as in our study, when estimating a multivariate probit model using the GHK simulation method for maximum likelihood estimation, the recommended number of replications (i.e. random draws) is equal to the square root of the sample size (thus, in Models 1 and 2, the number of draws is 16) (for another application, see e.g. Ziegler and Nogareda, 2009).

technological innovation (RQ1), cooperation with competitors, with HEIs and with public sector institutions are each significantly associated with greater probability of product innovation ( $p < 0.05$ ). Cooperation with competitors is beneficial, as it can lead to cost reduction (Belderbos *et al.*, 2004b), while universities can facilitate firms' product innovation given their broader knowledge base compared to other partners (Un *et al.*, 2010) and both HEIs and public sector institutions enable cooperation with low risk of knowledge leakage (Cassiman and Veuglers, 2002). In contrast, only cooperation with public sector institutions appears to increase the likelihood of undertaking process innovation ( $p < 0.05$ ), which is also consistent with the importance of concerns over knowledge leakage.

Our empirical results regarding the effects of vertical and horizontal cooperation on technological product and process innovations conflict with Tomlinson and Fai (2013), who report the largest and most significant impact of cooperation with suppliers among UK manufacturing SMEs with no effect of horizontal cooperation, but are in line with their reported insignificant impact of cooperation with customers on process innovation. This dissimilarity may reflect different country coverage; for example, our study includes data from seven EU regions, while Tomlinson and Fai (2013) focus solely on UK SMEs. Nonetheless, our findings are consistent with those of Nieto and Santamaria (2010), who observe that process innovations are less attractive for SMEs and, in line with this argument, found no significant impact of vertical cooperation on process innovation. For Spanish SMEs in low and medium-low tech sectors, Hervás-Oliver *et al.* (2011) found positive effects of cooperation with suppliers and universities on both product and process innovations, while no significant effects of cooperation with customers and competitors. These findings are partially consistent with ours.

Reviewing non-technological innovations (RQ2), cooperation with suppliers, private sector institutions and with public sector institutions each increase the probability of introducing organizational innovation, while no form of cooperation affects marketing innovation. These findings are partly in line with Pippel (2014), who reports a positive impact of cooperation with suppliers, consultants, other firms within an enterprise group and universities on both organizational and marketing innovations, while cooperation with customers only affects organizational innovation.

Overall, these results suggest that cooperation with public sector institutions is the only cooperative tie to affect all three of product, process and organizational innovations ( $p < 0.05$ ), but not marketing innovation ( $p > 0.05$ ). Although only 21 percent of SMEs in our sample cooperate with the public sector, which is in line with Mohnen and Hoareau (2003), who report that mostly large firms tend to cooperate with government labs and HEIs, we can see that this

type of cooperation increases not only product and process innovations but also organizational innovations. Finally, cooperation with customers does not appear to significantly impact innovation, irrespective of its type.

Concerning the control variables, firm size has a positive effect on organizational innovation ( $p < 0.05$ ), i.e. medium-sized firms are more likely to introduce this type of innovation than are smaller firms. Exporting activities have no effect on any type of innovation. In relation to our theoretical expectation this is anomalous (see above, “model specification”); however, in view of low statistical significance ( $p > 0.05$ ) of this estimate, we do not attempt interpretation. Very strong competitive pressure reduces the probability of introducing technological – product and process – innovations, but has no effect on non-technological innovations. These two estimates are each highly statistically significant ( $p < 0.01$ ) and consistent with the industrial organization prediction that high levels of competition adversely affect innovation.

With respect to the quasi fixed effects, an increase in the total resources dedicated to innovation is beneficial to introducing process, organizational and marketing innovations ( $p < 0.05$ ), but, rather surprisingly, has no effect on product innovation. In contrast, the most significant impact ( $p < 0.01$ ) on product innovation is found where established innovation capacity regarding this type of innovation exists. In other words, the probability of undertaking product innovation is associated with firms' established innovative capacity (initial conditions) for product innovation.

These findings are consistent with the resource-based view of the firm and the importance of absorptive capacity for firms within our sample. Yet our results also point to more subtle effects, whereby established capabilities may also exert negative effects on innovative outcomes: our results suggest that past innovation capacity in process innovation has an adverse effect on the current introduction of product innovation; and that established capacity for organizational innovation exerts a detrimental effect on the current introduction of process innovation. These negative influences from initial conditions or established innovation capacity in firms are consistent with "lock-in" effects (path dependency) (Teece, 1986) and suggest that SMEs in traditional manufacturing industries may experience considerable inertia in their processes and organization.

Model estimates with the breadth of cooperation as the variable of interest are shown in Model 2 (Table 3).<sup>7</sup> The impact and significance of the control variables is similar to those reported in Model 1 (Table 3). The results show that the breadth of cooperation (i.e. the number of cooperative relationships) is positively and significantly ( $p < 0.05$ ) associated with the probability of introducing all but marketing innovation. In addition, regarding RQ4, while our results in each case hint at a curvilinear relationship between the breadth of cooperation and technological and non-technological innovations (positive linear effects are consistently matched by the hypothesized negative quadratic effects), only linear effects are statistically significant ( $p < 0.05$ ) (similar findings when innovation output is measured by innovative sales are reported in Love *et al.* 2014). This finding suggests that the curvilinear relationship suggested in RQ4 may not apply to SMEs generally. The manufacturing SMEs in our sample benefit from having broad and extensive cooperative ties with different partners, but we do not find evidence that the positive innovation effects diminish and eventually reverse as the number of partnerships reaches a certain level (i.e. there is no turning point).

Table 4 reports the diagnostic statistics for Models 1 and 2. Each correlation coefficient  $\rho$  represents a pairwise correlation between the error terms of the four equations in each model. If the coefficient is statistically significant, that implies that the error terms are correlated and that the two equations should be estimated jointly (Greene, 2012: 747). In other words, a correlation coefficient measures the correlation between the outcomes after the observed heterogeneity (i.e. observed firm characteristics) is taken into account. Given that all the correlation coefficients are highly statistically significant, we conclude that multivariate probit is the appropriate model for our sample.

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 Insert Table 4 about here  
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The economic interpretation of these uniformly positive and highly significant ( $p < 0.01$ ) correlations between each pair of error terms is two-fold:

1. all four types of innovation have significant common unobserved factors; such that
2. if a positive change in an unobserved influence increases one type of innovation then, via positive correlations, it will increase the other three types also.

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<sup>7</sup> A separate model has to be estimated, because the breadth of cooperation is an exact linear combination of all seven cooperative partners.

This provides unambiguous evidence that all four types of innovation activities are complementary (Schmiedeberg, 2008). This complementarity is a contemporaneous effect – i.e. the unobserved influences act on all four types of innovation at the same time. Of course, this does not exclude the possibility of “lock-in” effects on one or more types of current innovations, from capabilities established in the past.

Table 5 shows the results of the ordered logit models for the dependent variable *Innovative sales* (Model 3 with individual cooperative partners and Model 4 with the breadth of cooperation and breadth squared as the variables of interest). The Model 3 estimates in relation to RQ3 suggest that cooperation with customers ( $p < 0.01$ ) and private sector institutions ( $p < 0.05$ ) positively and significantly increases innovative sales from product and process innovations. Therefore, while cooperation with customers was the only cooperative tie without any effect on technological and non-technological innovations (Table 3), it exerts the largest and a highly significant effect ( $p < 0.01$ ) on innovative sales, which measures the commercial success of technological product and process innovations (Love *et al.*, 2014). Miotti and Sachwald (2003) found that vertical cooperation in a sample of French firms, unlike cooperation with competitors and public institutions, was the only form of cooperation that increases innovative sales, whilst Von Hippel (1988) identified cooperation with customers as relevant for mitigating the risk inherent to the market introduction of innovation (Belderbos *et al.*, 2004b). Our findings on the impact of cooperation with customers on innovative sales are consistent with these previous contributions to the literature. In addition, a significant influence of cooperation with private sector institutions could be explained by both a low likelihood of knowledge leakage (i.e., no commercial risk), coupled with a broad knowledge base that this cooperative partner can provide to firms.

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Insert Table 5 about here  
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Overall, the findings reported in Model 3 coincide with those of Harris *et al.* (2000), who argue that cooperation for innovation is important in facilitating innovation activities, but does not necessarily result in commercial success. In other words, cooperative partners that influence the introduction of technological innovation do not significantly affect the commercial success of this form of innovation.

Although not all types of cooperation prove to be statistically significant promoters of commercial success, the Model 4 estimates imply that breadth of cooperation has a highly positive effect on innovative sales ( $p < 0.01$ ), without exhibiting a statistically significant non-linear relationship with this measure of innovation output. These results imply that the answer to RQ4, in the case of innovative sales as a measure of innovation performance, is the same as for technological and non-technological types of innovation. That is, there is no non-linear relationship between the breadth of cooperation and innovative sales. Therefore, looking at both Models 2 and 4, we conclude that SMEs benefit from diverse cooperative networks, which are reflected in higher innovation performance as well as in the commercial success of innovation. Given that most SMEs in our sample are innovative firms, this conclusion echoes that of Freel (2000), who argues that innovative small firms engage in diverse and extensive cooperation with a number of partners, although the impact of cooperation with each individual partner might not be necessarily beneficial to small firms.

Looking at the impact of other explanatory variables in Models 3 and 4, it can be noted that very strong competitive pressure is again negatively associated with innovative sales ( $p < 0.05$ ), while initial conditions with respect to firms' established innovation capacity for product innovation positively affect ( $p < 0.05$ ) innovation performance measured by innovative sales.

Table 6 presents the marginal effects for Model 3. These reveal striking results for the influence of three variables of interest on firms' abilities to achieve commercial success through innovation: above average or leading capacity for product innovation (*Capacity\_product*), cooperation with customers (*Coop\_customers*), and cooperation with private-sector institutions (*Coop\_private sector*) all substantially reduce the probabilities of firms being in the lower categories of innovative sales while increasing the probability of being in the higher categories. For example, cooperation with customers (*Coop\_customers*) reduces the probabilities of firms being in the three lower categories by, respectively, 4.2, 10.9 and 11.5 per cent, while increasing the probability of being in the three higher categories by, respectively, by 6.2, 10.1 and 12.2 per cent. For each of these three variables, in each of these categories of innovative sales – with the one exception of the effect of cooperation with private-sector institutions (*Coop\_private sector*) (for which  $p < 0.10$ ) – these estimates are uniformly statistically significant ( $p < 0.05$ ), while in no case is there a statistically significant effect for the median category of 11-15 per cent. In addition, the same pattern appears for above average or leading capacity for organizational innovation (*Capacity\_org*), although without the same levels of statistical significance in each category. Finally, these estimates also contribute to

understanding the effects of competition on the ability of firms to achieve commercial success through innovation: very high competitive pressures increase the probability of firms being in the lower categories while reducing the probability of being in the higher categories. Taken as a whole, the estimated effects are neither too large to be implausible nor too small to be economically irrelevant. The statistically significant estimates range from the effect of cooperation with private-sector institutions (*Coop\_private sector*) on the probability of a firm being in the lowest category of commercial success (a reduction of 2.2%) to the effect of cooperating with customers on the probability of being in the highest category of commercial success (an increase of 12.2%). These are economically substantial effects. In all respects, the marginal effects for Model 4 are similar (see Table 7). The one addition is the effect of breadth of cooperation on commercial success: an additional cooperative partner is associated with reductions of, respectively, 2.8, 6.9 and 6.2 per cent in the probabilities of a firm being in one of the three lower categories, and increases of, respectively, 4.5, 5.6 and 6.0 per cent in the probabilities of being in one of the three higher categories. (Once again, there is no statistically significant effect with respect to the median category).

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Insert Table 6 about here

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Insert Table 7 about here

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## 5. Conclusions

In this study we investigate how cooperation with different partners affects the innovation performance of SMEs in traditional manufacturing industries in the European Union. Innovation performance is measured in two ways: as the introduction of technological and non-technological innovations; and as innovative sales, with this latter measurement reflecting the commercial success of technological innovations. Additionally, we report the impact of breadth of cooperation on both measures of innovation performance.

Table 8 summarizes all of the estimated effects of cooperation reported in this study, by setting out the statistically significant effects of different types of cooperation on the different measures of innovation performance. Our study provides four substantive



conclusions. The first is that cooperation promotes innovation by SMEs in traditional manufacturing industry. This is demonstrated most clearly by the uniformly positive impact of additional partnerships (*Breadth*) on both the types of innovation enacted and on the commercial success of technological innovation: additional partners are associated with firms enacting higher levels of product, process and organizational innovation as well as with reduced probabilities of achieving lower levels of innovative sales and increased probabilities of achieving higher levels of innovative sales. Moreover, the estimated magnitudes (see Table 7) suggest that these estimated commercial effects are economically substantial.

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Insert Table 8 about here  
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Concerning RQ4, our estimates do not provide statistically significant support for the commonly observed non-linear (inverted-U) relationship between the breadth of cooperation and innovation performance. A little microeconomic theorizing may help to relate this finding to known specifics of SMEs in traditional manufacturing industries. We assume: (1) that for a representative firm the total innovation benefit (TIB) is a positive function of search effort (proxied by the number of cooperative external relationships) subject to diminishing returns (perhaps reflecting absorptive capacity as a fixed factor); (2) that total innovation costs (TIC) rise linearly in proportion to the number of cooperative external relationships; and (3) that for the first cooperative external relationship  $TIB > TIC$  (otherwise the optimum private number of cooperative external relationships is zero). The corollary is that the total innovation return (i.e. the difference between TIB and TIC) at first rises with each successive cooperative external relationship and then falls. Eventually the marginal innovation return becomes negative once the “oversearch” threshold level of search – at which  $TIB = TIC$  – is exceeded. In the light of this reasoning, our findings have implications for both business and public policy. First, because cooperation is less well established among SMEs in traditional manufacturing industries than among firms more generally (the mean number is 1.6 in our sample), the number of partnerships is starting from a low base and thus the innovation effects are less subject to diminishing returns. If so, then the level of cooperation among traditional sector SMEs is not only low in a numerical sense but also in the economic sense that such firms typically have not yet reached a level of search that is optimal from the perspective of innovation. Secondly, policy makers need have no fear that policies designed to induce marginal increases in external

cooperation by traditional sector SMEs will push them towards “oversearch” from either a private or social perspective.

These findings emphasize the importance of diverse and extensive cooperative networks for European SMEs in traditional manufacturing industry. For owners and managers, the emergent message appears to be that innovation performance can be enhanced if a portfolio approach to cooperation is adopted. This approach to cooperation promotes both innovation and its commercialization.

The second conclusion is that among individual types of cooperation the performance effects are heterogeneous. First, in relation to RQ1 on the impact of cooperative relationships on technological innovation, while cooperation with competitors, with HEIs and with public sector institutions increase the likelihood of product innovation, only cooperation with public sector institutions is associated with greater probability of process innovation. Regarding RQ2 on the influence of various cooperative partners on non-technological innovation, the likelihood of organizational innovation increases when firms cooperate with suppliers, private sector institutions and with public sector institutions. However, marketing innovation appears to be not affected by any type of cooperation. In summing up, most of the estimated positive effects (four from seven) arise from cooperation with other public-sector knowledge providers (three) or HEIs (one). This is consistent with public support measures designed to promote partnerships between SMEs and external knowledge providers (through for example, “innovation vouchers”, see e.g. Matulova et al., 2015). Second, regarding RQ3 on the impact of cooperation on innovative sales, our estimates consistently indicate that cooperation with customers and private-sector knowledge providers promote technological innovation with commercial impact, but do not provide evidence for such positive performance effects from other types of partner.

In spite of our particular focus on SMEs in traditional industries, our findings on the innovation effects of particular forms of cooperation are broadly in line with studies using less restrictive samples. First, in common with Miotti and Sachwald (2003) we find that cooperation with customers has a highly positive impact on innovative sales; although, contrary to other studies (e.g. Nieto and Santamaria 2010; Lasagni 2012; Tomlinson and Fai 2013), we found that vertical cooperation with customers and suppliers has no systematic impact on product and process innovations. Second, our finding of a positive effect of horizontal cooperation with competitors on product innovation but not on process innovation is, in part, consistent with Tomlinson and Fai (2013), who report an insignificant effect of cooperation with competitors on both forms of technological innovation. Third, our finding that cooperation with public

sector knowledge providers is positively associated with product, process and organizational innovations is consistent with Lasagni (2012), who found a positive influence of cooperation with research organizations on product innovation. Moreover, our finding that cooperation with public sector knowledge providers does not enhance the commercial success of technological innovations is consistent with Zeng et al. (2010). Finally, concerning non-technological innovations, our findings partly coincide with Sánchez-González (2014), who reports a positive impact of each cooperative partner (suppliers, customers, competitors, experts, and universities), but are more in line with Pippel (2014), who found heterogeneous performance effects of different cooperative partners on non-technological innovations.

The third conclusion extends our discussion of heterogeneity. Namely, cooperative partners that influence the introduction of technological innovation do not necessarily affect the commercial success of this form of innovation; and vice versa. For example, cooperation with competitors might promote product innovation, but this is not reflected in its commercial impact. In contrast, while cooperation with customers was not found to affect technological innovations, it exerts the largest and a highly significant effect on innovative sales.

The fourth conclusion arises from the finding that all four types of innovation have significant common unobserved factors. Accordingly, if a positive change in an unobserved influence at firm level (e.g. a change in management) increases one type of innovation then it will increase the other three types as well. This provides unambiguous evidence that all four types of innovation activities are complementary. For policy makers this suggests that public support programs to promote SME innovation in traditional manufacturing industry should be demand-led (i.e. flexible with respect to SME needs) rather than supply led (i.e. narrowly prescriptive with respect to one or other aspect of technological or non-technological innovation). Correspondingly, owners and managers are best advised to take a holistic approach to innovation (i.e. to be aware that innovation in one area may well require complementary innovations elsewhere).

As well as new findings for our variables of interest, the estimated effects of the control variables are either consistent with the existing literature (e.g. on the effects of competition and absorptive capacity) or suggest further lines of enquiry (e.g. with respect to the “lock in” effects of established innovative capacities). We find that very high levels of competitive pressure tend to reduce firms' innovativeness, which is in line with the Industrial Organization literature. We also find that established absorptive capacity can have both positive and negative impacts, depending on the type of innovation. This finding might be relevant for owners and managers, as it may indicate potential for adverse "lock-in" effects.

We recognize some inherent limitations to our study. First, the survey questionnaire did not contain a question on the intensity of cooperative ties, which would have enabled exploration of the innovation effects of depth of cooperation. Second, although, within the limitations of cross-section survey data, we do control for firms' time-invariant (or slowly moving) characteristics, panel data with at least four or five waves would be required to explore the medium- and long-run effects of cooperation for innovation (Belderbos *et al.*, 2004b; Pittaway *et al.*, 2004; Aschhoff and Schmidt, 2008). Third, our data does not contain information on the duration of collaboration, which means that we cannot assign different weights reflecting the length of collaboration. Moreover, we do not have information about the number of specific partners at different times during the survey period, e.g. whether the firm collaborated with one or three suppliers at the same time or at different times in the surveyed period. Finally, the GPrix survey was conducted during the global financial crisis. Future research might investigate how cooperation influences innovation in traditional manufacturing SMEs in a more recent period.

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**Table 1. Descriptive statistics**

<b>Variables</b>	<b>Mean</b>	<b>Standard deviation</b>	<b>Minimum</b>	<b>Maximum</b>
<i>Product innovation</i>	0.811	0.399	0	1
<i>Process innovation</i>	0.827	0.379	0	1
<i>Organizational innovation</i>	0.681	0.467	0	1
<i>Marketing innovation</i>	0.610	0.489	0	1
<i>Innovative sales</i>	4.180	1.924	1	7
<i>Size</i>	35.563	45.205	0	230
<i>Competition</i>	0.232	0.423	0	1
<i>Export</i>	19.858	30.239	0	100
<i>Resources</i>	0.362	0.482	0	1
<i>Capacity_product</i>	0.264	0.442	0	1
<i>Capacity_process</i>	0.209	0.407	0	1
<i>Capacity_org</i>	0.134	0.341	0	1
<i>Capacity_marketing</i>	0.165	0.372	0	1
<i>Coop_within_group</i>	0.122	0.328	0	1
<i>Coop_suppliers</i>	0.323	0.468	0	1
<i>Coop_customers</i>	0.335	0.473	0	1
<i>Coop_competitors</i>	0.087	0.282	0	1
<i>Coop_private sector</i>	0.236	0.426	0	1
<i>Coop_HEIs</i>	0.307	0.462	0	1
<i>Coop_public sector</i>	0.205	0.404	0	1
<i>Breadth</i>	1.614	1.512	0	6
<i>Leather industry</i>	0.043	0.204	0	1
<i>Ceramic industry</i>	0.075	0.264	0	1
<i>Textile industry</i>	0.118	0.323	0	1
<i>Mechanical/metallurgy industry</i>	0.295	0.457	0	1
<i>Automotive industry</i>	0.106	0.309	0	1

<i>Food processing industry</i>	0.169	0.376	0	1
<i>Other manufacturing industries</i>	0.193	0.395	0	1
<i>Spain</i>	0.193	0.395	0	1
<i>France</i>	0.094	0.293	0	1
<i>Germany</i>	0.110	0.314	0	1
<i>Italy</i>	0.165	0.372	0	1
<i>Netherlands</i>	0.102	0.304	0	1
<i>Portugal</i>	0.055	0.229	0	1
<i>United Kingdom</i>	0.280	0.450	0	1

**Table 2. Correlation matrix**

Independent variables	1.	2.	3.	4.	5.	6.	7.	8.	9.	10.	11.	12.	13.	14.	15.	16.
<i>1. Size</i>	1.000															
<i>2. Competition</i>	0.092	1.000														
<i>3. Export</i>	0.267***	-0.115**	1.000													
<i>4. Resources</i>	0.052	-0.103*	0.087	1.000												
<i>5. Capacity_product</i>	0.018	-0.085	0.107*	-0.033	1.000											
<i>6. Capacity_process</i>	0.072	0.029	-0.067	-0.064	0.529***	1.000										
<i>7. Capacity_org</i>	0.037	-0.050	0.036	0.014	0.435***	0.437***	1.000									
<i>8. Capacity_marketing</i>	0.038	-0.111*	-0.084	-0.004	0.400***	0.583***	0.456***	1.000								
<i>9. Coop_within_group</i>	0.238***	0.041	-0.041	-0.044	0.056	0.155***	0.065	0.173***	1.000							
<i>10. Coop_suppliers</i>	0.071	-0.126**	0.045	0.108*	0.099*	0.044	0.079	0.084	0.228***	1.000						
<i>11. Coop_customers</i>	-0.022	-0.056	-0.035	0.102*	0.054	0.129**	0.015	0.157***	0.199***	0.414***	1.000					
<i>12. Coop_competitors</i>	-0.074	-0.079	-0.105*	0.092	-0.065	-0.004	-0.058	0.018	0.092	0.067	0.193***	1.000				
<i>13. Coop_private_sector</i>	0.106*	-0.101*	0.150***	0.018	0.139**	0.160***	0.190***	0.200***	0.089	0.220***	0.128**	0.069	1.000			
<i>14. Coop_HELs</i>	0.016	-0.103*	0.133**	0.096*	0.054	0.123**	0.058	0.183***	0.077	0.147***	0.206***	0.095*	0.334***	1.000		
<i>15. Coop_public sector</i>	0.061	-0.069	0.138**	0.124**	0.086	0.049	0.064	0.138**	0.016	0.062	-0.032	0.035	0.192***	0.342***	1.000	
<i>16. Breadth</i>	0.102*	0.091	-0.140**	0.141**	0.127**	0.183***	0.120**	0.265***	0.419***	0.621***	0.608***	0.354***	0.564***	0.627***	0.439***	1.000

Notes: \*\*\*p<0.01; \*\*p<0.05; \* p<0.10.

**Table 3. Multivariate probit model: dependent variables *Product innovation, Process innovation, Organizational innovation, and Marketing innovation***

Independent variables	Model 1				Model 2			
	Product innovation	Process Innovation	Organizational Innovation	Marketing innovation	Product innovation	Process innovation	Organizational innovation	Marketing innovation
<i>Size</i>	0.000 (0.003)	0.004 (0.003)	0.008** (0.003)	-0.001 (0.002)	-0.000 (0.002)	0.004 (0.003)	0.007** (0.003)	-0.001 (0.002)
<i>Export</i>	0.000 (0.004)	-0.008 <sup>†</sup> (0.004)	0.002 (0.004)	0.003 (0.003)	-0.000 (0.004)	-0.007 <sup>†</sup> (0.004)	0.003 (0.003)	0.002 (0.003)
<i>Competition</i>	-1.068*** (0.251)	-0.675*** (0.252)	-0.188 (0.245)	-0.354 (0.220)	-0.963*** (0.247)	-0.699*** (0.241)	-0.242 (0.237)	-0.341 (0.219)
<i>Resources</i>	0.203 (0.212)	0.798*** (0.256)	0.505** (0.208)	0.520*** (0.201)	0.375* (0.220)	0.861*** (0.262)	0.559*** (0.199)	0.456** (0.195)
<i>Capacity_product</i>	1.290*** (0.340)	0.282 (0.318)	-0.042 (0.278)	0.518 <sup>†</sup> (0.289)	1.217*** (0.355)	0.304 (0.314)	0.016 (0.262)	0.438 (0.285)
<i>Capacity_process</i>	-1.057*** (0.331)	0.477 (0.378)	0.515 (0.324)	-0.394 (0.299)	-0.934*** (0.330)	0.393 (0.361)	0.440 (0.330)	-0.346 (0.300)
<i>Capacity_org</i>	0.419 (0.423)	-0.845** (0.391)	-0.446 (0.367)	0.590 (0.366)	0.224 (0.379)	-0.854** (0.382)	-0.317 (0.340)	0.619 <sup>†</sup> (0.352)
<i>Capacity_marketing</i>	0.407 (0.399)	0.507 (0.378)	0.261 (0.350)	0.214 (0.331)	0.337 (0.417)	0.593 (0.380)	0.196 (0.339)	0.213 (0.327)
<i>Coop_within group</i>	0.308 (0.320)	-0.031 (0.380)	-0.067 (0.370)	0.554 <sup>†</sup> (0.291)				
<i>Coop_suppliers</i>	0.203 (0.216)	0.266 (0.288)	0.655*** (0.227)	0.171 (0.239)				
<i>Coop_customers</i>	0.176 (0.216)	0.305 (0.307)	0.319 (0.240)	0.286 (0.249)				
<i>Coop_competitors</i>	1.101** (0.532)	0.954 (0.612)	0.199 (0.356)	0.260 (0.334)				
<i>Coop_private sector</i>	0.188 (0.336)	0.342 (0.271)	0.618** (0.284)	0.302 (0.235)				
<i>Coop_HEIs</i>	0.648** (0.312)	0.102 (0.262)	0.222 (0.252)	0.139 (0.227)				
<i>Coop_public sector</i>	1.209** (0.585)	0.749** (0.336)	0.748** (0.309)	-0.029 (0.281)				
<i>Breadth</i>					0.429**	0.404**	0.464***	0.258

<i>Breadth_square</i>					(0.195)	(0.182)	(0.176)	(0.161)
					-0.010	-0.025	-0.013	-0.011
					(0.045)	(0.041)	(0.042)	(0.036)
<i>Constant</i>	0.081	0.223	-0.627**	-0.631**	0.015	0.033	-0.788**	-0.569 <sup>†</sup>
	(0.350)	(0.370)	(0.315)	(0.317)	(0.323)	(0.359)	(0.313)	(0.307)
Industry DVs	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Country DVs	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
No of obs.	254				254			
Log pseudolikelihood	-381.20				-391.28			
Wald $\chi^2$ (108)	517.88***				400.96***			
LR test on $\rho_{21}=\rho_{31}=\rho_{41}=\rho_{32}=\rho_{42}=\rho_{43}=0$ ; $\chi^2$ (6)=75.09***								LR test on $\rho_{21}=\rho_{31}=\rho_{41}=\rho_{32}=\rho_{42}=\rho_{43}=0$ ; $\chi^2$ (6)=73.58***

Notes: Robust standard errors in parentheses. The number of draws is 16. \*\*\* p<0.01, \*\* p<0.05, <sup>†</sup> p<0.10.

**Table 4. Correlation coefficients for Models 1 and 2**

Correlation coefficients	Model 1	Model 2
$\rho_{21}$	0.792*** (0.081)	0.762*** (0.076)
$\rho_{31}$	0.518*** (0.114)	0.488*** (0.124)
$\rho_{41}$	0.556*** (0.096)	0.522*** (0.101)
$\rho_{32}$	0.560*** (0.157)	0.582*** (0.163)
$\rho_{42}$	0.549*** (0.142)	0.523*** (0.125)
$\rho_{43}$	0.552*** (0.103)	0.499*** (0.110)

Notes: \*\*\* p<0.01;  $\rho_{21}$  denotes the correlation coefficient between the error terms of two equations *Process innovation and Product innovation*;  $\rho_{31}$  denotes the correlation coefficient between the error terms of equations *Organizational innovation and Product innovation*;  $\rho_{41}$  denotes the correlation coefficient between the error terms of equations *Marketing innovation and Product innovation*;  $\rho_{32}$  denotes the correlation coefficient between the error terms of equations *Organizational innovation and Process innovation*;  $\rho_{42}$  denotes the correlation coefficient between the error terms of equations *Marketing innovation and Process innovation*;  $\rho_{43}$  denotes the correlation coefficient between the error terms of equations *Marketing innovation and Organizational innovation*.

**Table 5. Ordered logit model: dependent variable - innovative sales**

<b>Independent variables</b>	<b>Model 3</b>	<b>Model 4</b>
<i>Size</i>	-0.002 (0.002)	-0.003 (0.002)
<i>Export</i>	0.002 (0.004)	0.003 (0.004)
<i>Competition</i>	-0.706** (0.336)	-0.668** (0.340)
<i>Resources</i>	0.518 <sup>†</sup> (0.285)	0.473 <sup>†</sup> (0.270)
<i>Capacity_product</i>	0.867** (0.351)	0.979*** (0.327)
<i>Capacity_process</i>	0.518 (0.369)	0.502 (0.350)
<i>Capacity_org</i>	0.805 (0.502)	0.743 (0.469)
<i>Capacity_marketing</i>	-0.362 (0.403)	-0.335 (0.379)
<i>Coop_within group</i>	0.047 (0.377)	
<i>Coop_suppliers</i>	0.364 (0.318)	
<i>Coop_customers</i>	1.176*** (0.334)	
<i>Coop_competitors</i>	-0.404 (0.499)	
<i>Coop_private sector</i>	0.616** (0.286)	
<i>Coop_HEIs</i>	0.474 <sup>†</sup> (0.269)	
<i>Coop_public sector</i>	0.309 (0.334)	
<i>Breadth</i>		0.657*** (0.231)
<i>Breadth_square</i>		-0.033 (0.051)
<i>Constant1</i>	-2.231*** (0.501)	-2.152*** (0.494)
<i>Constant2</i>	-0.672 (0.444)	-0.596 (0.445)
<i>Constant3</i>	0.509 (0.446)	0.553 (0.451)
<i>Constant4</i>	1.149*** (0.445)	1.172*** (0.452)
<i>Constant5</i>	2.128*** (0.440)	2.119*** (0.439)
<i>Constant6</i>	3.118*** (0.458)	3.089*** (0.446)
Industry DVs	Yes	Yes
Country DVs	Yes	Yes
No of obs.	261	261
McFadden pseudo R <sup>2</sup>	0.124	0.114
Log pseudolikelihood	-438.98	-443.93
LR $\chi^2$	$\chi^2 (27) = 126.71$ ***	$\chi^2 (22) = 107.69$ ***

Notes: Robust standard errors in parentheses. \*\*\* p<0.01, \*\* p<0.05, <sup>†</sup> p<0.10.

**Table 6. Marginal effects for Model 3**

<b>Independent variables</b>	<b>Outcome 1 Innovative sales 0%</b>	<b>Outcome 2 Innovative sales 1-5%</b>	<b>Outcome 3 Innovative sales 6-10%</b>	<b>Outcome 4 Innovative sales 11-15%</b>	<b>Outcome 5 Innovative sales 16-25%</b>	<b>Outcome 6 Innovative sales 26-50%</b>	<b>Outcome 7 Innovative sales &gt;50%</b>
<i>Size</i>	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)
<i>Export</i>	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)
<i>Competition</i>	0.035 <sup>†</sup> (0.020)	0.080 <sup>†</sup> (0.044)	0.059** (0.024)	-0.007 (0.012)	-0.056 <sup>†</sup> (0.030)	-0.057** (0.025)	-0.053** (0.023)
<i>Resources</i>	-0.020 <sup>†</sup> (0.011)	-0.051 <sup>†</sup> (0.028)	-0.051 <sup>†</sup> (0.029)	-0.005 (0.006)	0.034 <sup>†</sup> (0.019)	0.045 <sup>†</sup> (0.026)	0.048 <sup>†</sup> (0.028)
<i>Capacity_product</i>	-0.030** (0.012)	-0.079*** (0.031)	-0.088** (0.037)	-0.016 (0.014)	0.047*** (0.016)	0.076** (0.033)	0.090** (0.045)
<i>Capacity_process</i>	-0.018 (0.012)	-0.049 (0.032)	-0.053 (0.039)	-0.008 (0.011)	0.031 (0.020)	0.046 (0.034)	0.051 (0.040)
<i>Capacity_org</i>	-0.025 <sup>†</sup> (0.013)	-0.069** (0.035)	-0.083 (0.053)	-0.020 (0.024)	0.037*** (0.013)	0.072 (0.045)	0.090 (0.071)
<i>Capacity_marketing</i>	0.016 (0.021)	0.040 (0.047)	0.033 (0.033)	-0.002 (0.007)	-0.028 (0.034)	-0.030 (0.032)	-0.029 (0.029)
<i>Coop_within group</i>	-0.002 (0.015)	-0.005 (0.038)	-0.005 (0.037)	-0.000 (0.003)	0.003 (0.026)	0.004 (0.033)	0.004 (0.034)
<i>Coop_suppliers</i>	-0.014 (0.012)	-0.036 (0.031)	-0.036 (0.033)	-0.003 (0.006)	0.024 (0.020)	0.032 (0.029)	0.034 (0.032)
<i>Coop_customers</i>	-0.042*** (0.015)	-0.109*** (0.031)	-0.115*** (0.035)	-0.020 (0.014)	0.062*** (0.020)	0.101*** (0.031)	0.122*** (0.042)
<i>Coop_competitors</i>	0.019 (0.028)	0.045 (0.060)	0.035 (0.037)	-0.003 (0.012)	-0.032 (0.044)	-0.033 (0.038)	-0.031 (0.034)
<i>Coop_private sector</i>	-0.022** (0.011)	-0.058** (0.026)	-0.063** (0.030)	-0.010 (0.010)	0.036** (0.015)	0.055** (0.026)	0.062 <sup>†</sup> (0.035)
<i>Coop_HEIs</i>	-0.018 <sup>†</sup> (0.010)	-0.046 <sup>†</sup> (0.026)	-0.048 <sup>†</sup> (0.028)	-0.005 (0.007)	0.031 <sup>†</sup> (0.017)	0.042 (0.025)	0.045 (0.027)
<i>Coop_public sector</i>	-0.012 (0.012)	-0.030 (0.032)	-0.031 (0.035)	-0.004 (0.007)	0.020 (0.020)	0.027 (0.030)	0.029 (0.034)

Notes: Robust standard errors in parentheses; \*\*\* p<0.01, \*\* p<0.05, <sup>†</sup> p<0.10; Industry and country DVs included.



**Table 7. Marginal effects for Model 4**

Independent variables	Outcome 1	Outcome 2	Outcome 3	Outcome 4	Outcome 5	Outcome 6	Outcome 7
	Innovative sales 0%	Innovative sales 1-5%	Innovative sales 6-10%	Innovative sales 11-15%	Innovative sales 16-25%	Innovative sales 26-50%	Innovative sales >50%
<i>Size</i>	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)
<i>Export</i>	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)
<i>Competition</i>	0.034 (0.021)	0.077 <sup>†</sup> (0.044)	0.054** (0.023)	-0.006 (0.011)	-0.051 <sup>†</sup> (0.029)	-0.054** (0.025)	-0.053** (0.025)
<i>Resources</i>	-0.019 <sup>†</sup> (0.011)	-0.048 <sup>†</sup> (0.028)	-0.045 <sup>†</sup> (0.027)	-0.004 (0.006)	0.030 <sup>†</sup> (0.017)	0.041 <sup>†</sup> (0.024)	0.046 (0.028)
<i>Capacity_product</i>	-0.035*** (0.012)	-0.091*** (0.030)	-0.096*** (0.034)	-0.019 (0.013)	0.047*** (0.015)	0.085*** (0.030)	0.109** (0.045)
<i>Capacity_process</i>	-0.019 (0.012)	-0.049 (0.032)	-0.050 (0.036)	-0.007 (0.010)	0.029 (0.018)	0.044 (0.032)	0.052 (0.039)
<i>Capacity_org</i>	-0.025 <sup>†</sup> (0.013)	-0.067 <sup>†</sup> (0.035)	-0.075 (0.049)	-0.017 (0.020)	0.033*** (0.012)	0.065 (0.041)	0.085 (0.067)
<i>Capacity_marketing</i>	0.016 (0.020)	0.037 (0.045)	0.029 (0.030)	-0.001 (0.006)	-0.025 (0.031)	-0.028 (0.031)	-0.028 (0.029)
<i>Breadth</i>	-0.028** (0.012)	-0.069*** (0.026)	-0.062*** (0.023)	-0.003 (0.007)	0.045** (0.018)	0.056*** (0.021)	0.060*** (0.023)
<i>Breadth_sq</i>	0.001 (0.002)	0.003 (0.005)	0.003 (0.005)	0.000 (0.000)	-0.002 (0.004)	-0.003 (0.004)	-0.003 (0.005)

Notes: Robust standard errors in parentheses; \*\*\* p<0.01, \*\* p<0.05, <sup>†</sup> p<0.10; Industry and country DVs included.

**Table 8. Effects of cooperation partnerships on innovation outcomes**

Types of cooperative partnership	Types of innovation				Commercial impact of technological innovation							
	Techno		Non-techno		Change	Change in innovative sales by category						
	Prod	Proc	Org.	Mkt.		Inn. sales	0%	1-5%	6-10%	11-15%	16-25%	26-50%
<i>Coop_within group</i>												
<i>Coop_suppliers</i>			+									
<i>Coop_customers</i>					+	-	-	-		+	+	+
<i>Coop_competitors</i>	+											
<i>Coop_private sector</i>			+		+	-	-	-		+	+	
<i>Coop_HEIs</i>	+											
<i>Coop_public sector</i>	+	+	+									
<i>Breadth</i>	+	+	+		+	-	-	-		+	+	+

Notes: + denotes a statistically significant positive effect (p<0.05); - a statistically significant negative effect (p<0.05); and blank indicates no statistically significant effect.

Source: Types of innovation – Table 3; Change in innovative sales – Table 5; and Change in innovative sales by category – Tables 6 and 7.